

# Speech modeling and automatic speech recognition

Lexicons and language models  
for ASR using pocketsphinx  
- experiments -

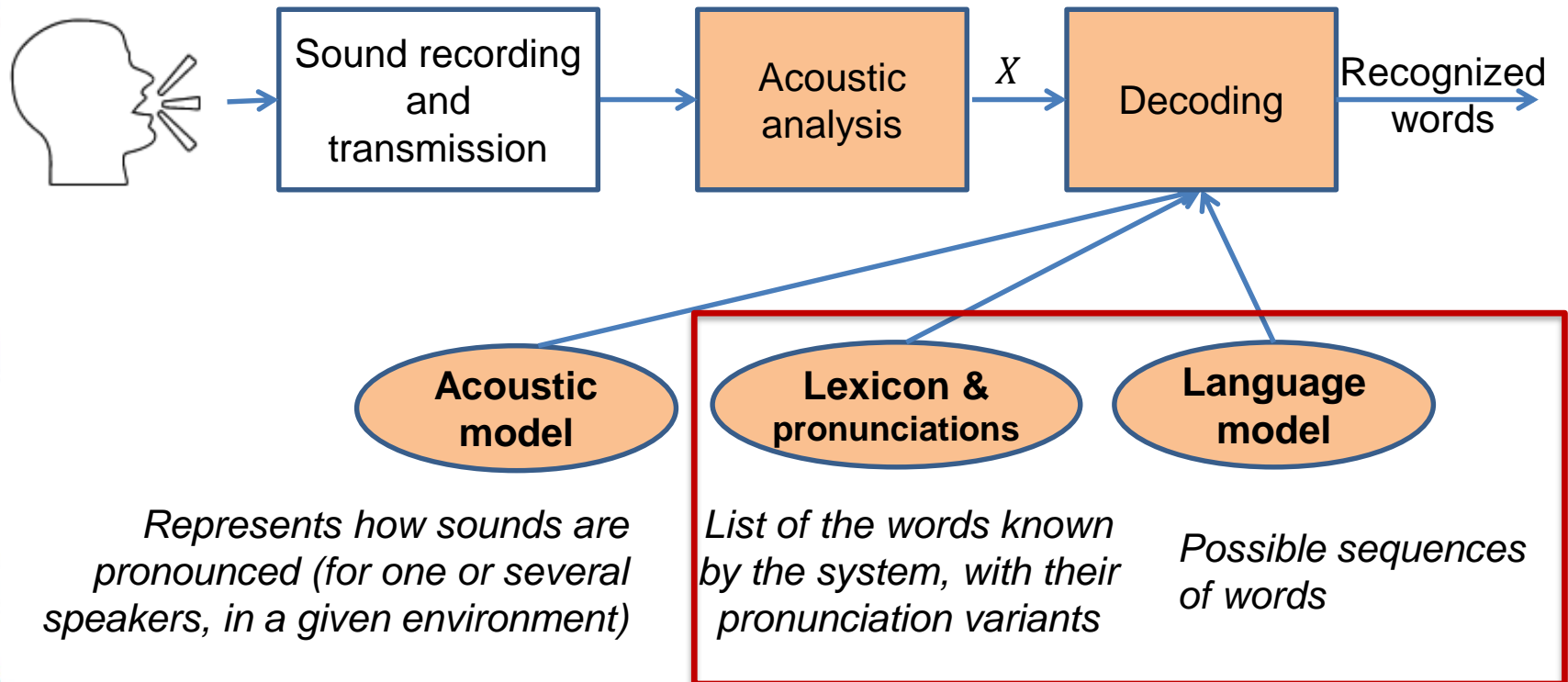


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# Automatic speech recognition



# Lexicon

- Specify a list of words and their pronunciation variants
- Example (excerpt from pocketsphinx lexicon)

one	HH W AH N
one (2)	W AH N
person	P ER S AH N
quarter	K AO R T ER
quarter (2)	K W AO R T ER
quarters	K W AO R T ER Z
quit	K W IH T

word

pronunciation

numbering  
of pronunciation  
variants

# Language models

- Provide information on the possible sequences of words
- **Context-free grammars**
  - Specify sequences of words corresponding to « sentences » (i.e. global constraints)
- **n-gram statistical model**
  - Provide local constraints (on sequences of  $n$  words)
  - Estimation from text corpora
- **Neural network models**
  - Many different approaches proposed in the literature

# Context-free grammars

- Rules (or graphs) that describe exactly the allowed sentences (sequences of words) of the language (e.g., isolated digits, numbers, ...)

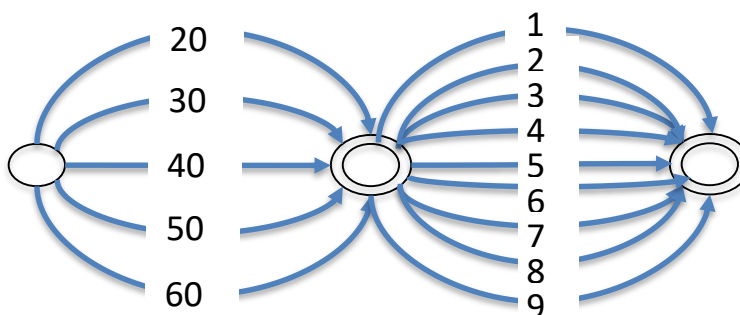
– Ex. Number\_from\_20\_to\_69

Number\_from\_20\_to\_69 = Tens\_from\_20\_to\_60

| Tens\_from\_20\_to\_60 . Units\_from\_1\_to\_9 ;

Tens\_from\_20\_to\_60 = 20 | 30 | 40 | 50 | 60 ;

Units\_from\_1\_to\_9 = 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 ;



- Global constraints on the sentences
- Complex and difficult to define for large vocabularies
- Do not allow the recognition of sentences that do not respect the grammar

# jsgf ⇔ JSpeech Grammar Format

- The JSpeech Grammar Format (JSGF) is a *platform-independent, vendor-independent textual representation* of grammars for use in speech recognition

- Example of JSGF grammar

```
#JSGF V1.0;
```

```
/**
```

```
 * JSGF Grammar for Turtle example
```

```
 */
```

```
grammar goforward;
```

```
public <move> = go forward ten meters;
```

```
public <move2> = go <direction> <distance> [meter | meters];
```

```
<direction> = forward | backward;
```

```
<distance> = one | two | three | four | five | six | seven |  
eight | nine | ten;
```

# N-gram language model

- Statistical model
- Probability of a word sequence

$$P(w_1, \dots, w_N) = P(w_1) \prod_{i=2..N} P(w_i | w_1, \dots, w_{i-1})$$

- N-gram approximation, as n-grams deal with **sequences of  $n$  words**

$$P(w_i | w_1, \dots, w_{i-1}) \triangleq P(w_i | w_{i-(n-1)}, \dots, w_{i-1})$$

- N-gram parameters are computed from large text corpora

# Examples of n-grams

- Sentence : *the sky is blue*

→ *<s> the sky is blue </s>*

including start and end of sentence symbols

- Unigrams

$$P(\text{Sentence}) = P(\textit{the}) \cdot P(\textit{sky}) \cdot P(\textit{is}) \cdot P(\textit{blue})$$

- Bigrams

$$P(\text{Sentence}) = P(\textit{the}|\textit{<s>}) \cdot P(\textit{sky}|\textit{the}) \cdot P(\textit{is}|\textit{sky}) \cdot P(\textit{blue}|\textit{is}) \cdot P(\textit{</s>}|\textit{blue})$$

$$P(\text{Sentence}) = P(w_1 | \textit{< s >}) \prod_{i=2, \dots, N} P(w_i | w_{i-1})$$

- Trigrams

$$P(\text{Sentence}) = P(\textit{the}|\textit{<s>}) \cdot P(\textit{sky}|\textit{<s> the}) \cdot P(\textit{is}|\textit{the sky}) \cdot P(\textit{blue}|\textit{sky is}) \cdot P(\textit{</s>}|\textit{is blue})$$



# Pocketsphinx

- *PocketSphinx is a lightweight speech recognition engine, specifically tuned for handheld and mobile devices, though it works equally well on the desktop*
- <https://github.com/cmusphinx/pocketsphinx>
- Available as a python package – thus easy to install and to use
- Remark:
  - The speech recognition performance depends on the quality of acoustic models (and on the adequation of the model with the test data)

# Lexicons and language models experiments

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- Lexicon and language model experiments
  - Pocketsphinx generic lexicon and language model (i.e., ngram)
  - Digit loop grammar (**to write**)
  - Grammar for sequences of digits of known length (**to write**)
- Speech corpus for performance evaluation
  - Sequences of digits (1 digit, 3 digits & 5 digits)
  - With added noise: signal to noise ratio (SNR) of 35 dB, 25 dB, 15 dB and 05 dB
- **Performance evaluation to do** – estimate word error rate with respect to
  - Language model
  - Length of digit sequence
  - Speaker group
  - Signal to noise ratio

# Digit corpus

In file « `td_corpus_digits.zip` »

`td_corpus_digits`

`SNR35dB`

`man`

`seq1digit_200_files`

→ files \*.wav & \*.ref of isolated digits

`seq3digits_100_files`

→ files \*.wav & \*.ref of 3 digit sequences

`seq5digits_100_files`

→ files \*.wav & \*.ref of 5 digit sequences

`woman`

*Same organization as for « man »*

`boy`

*Same organization as for « man »*

`girl`

*Same organization as for « man »*

`SNR25dB`

*Only « man » data; isolated digits, and 3 and 5 digit sequences*

`SNR15dB`

*Only « man » data; isolated digits, and 3 and 5 digit sequences*

`SNR05dB`

*Only « man » data; isolated digits, and 3 and 5 digit sequences*

# Lexicons & language models

- In file « `ps_data.zip` »

`ps_data`

`model`

`en_us`

→ English acoustic model

`lex`

`cmudict-en-us.dict`

→ English generic lexicon

`turtle.dic`

→ Small lexicon (for the jsfg grammar)

`lm`

`en-us.lm.bin`

→ Generic english Ngram model

`jsfg`

`goforward.gram`

→ Example of jsfg grammar

`exemple`

`goforward.raw`

→ Speech file

# Examples of usage of pocketsphinx

In file « `ps_exemples.zip` »

`ps_exemples`

`decoder_ngram.py` ([https://raw.githubusercontent.com/cmusphinx/pocketsphinx/master/swig/python/test/decoder\\_test.py](https://raw.githubusercontent.com/cmusphinx/pocketsphinx/master/swig/python/test/decoder_test.py))

From web – generic English lexicon and language model

Slightly modified to use lexicon and models from directory `ps_data`

Send speech data to decoder in small blocks

`decoder_jsgf.py` ([https://raw.githubusercontent.com/cmusphinx/pocketsphinx/master/swig/python/test/jsgf\\_test.py](https://raw.githubusercontent.com/cmusphinx/pocketsphinx/master/swig/python/test/jsgf_test.py))

From web – use a jsgf grammar

Slightly modified to use lexicon and models from directory `ps_data`

Send speech data to decoder in small blocks

`decoder_utt_ngram.py`

Modified version of `decoder_ngram.py` that sends the whole speech file in a single call

`decoder_utt_jsgf.py`

Modified version of `decoder_jsgf.py` that sends the whole speech file in a single call

# Setup

Here, assume running in a python virtual environment

```
python3 -m venv asr-env  
source asr-env/bin/activate
```

→ Create python virtual environment

→ Activate virtual environment

```
pip install pocketsphinx
```

→ Install pocketsphinx

```
python ps_exemples/decoder_ngram.py
```

→ Check that program runs

```
python ps_exemples/decoder_utt_ngram.py
```

→ Check that program runs

```
python ps_exemples/decoder_jsgf.py
```

→ Check that program runs

```
python ps_exemples/decoder_utt_jsgf.py
```

→ Check that program runs

```
pip install asr-evaluation
```

```
(usage: wer -i toto.ref toto.hyp)
```

→ To compute word error rate

.....

→ Run the experiments...

```
Deactivate
```

→ Exit the virtual environment

# Example of code using pocketsphinx decoder\_ngram.py

```
# Create a decoder with certain model
config = Decoder.default_config()
config.set_string('-hmm', 'ps_data/model/en-us')
config.set_string('-lm', 'ps_data/lm/en-us.lm.bin')
config.set_string('-dict', 'ps_data/lex/cmudict-en-us.dict')

# Decode streaming data.
decoder = Decoder(config)

decoder.start_utt()
stream = open('ps_data/exemple/goforward.raw', 'rb')
while True:
    buf = stream.read(1024)
    if buf:
        decoder.process_raw(buf, False, False)
    else:
        break
decoder.end_utt()

hypothesis = decoder.hyp()
print ('Best hypothesis: ', hypothesis.hypstr)
```

# Computation of word error rate

- Need two files
  - `Data.ref` ⇔ for the reference (i.e., transcriptions of the speech data)
  - `Data.hyp` ⇔ for the ASR hypotheses (i.e., speech recognition output)
- Warning – **files must be aligned**:
  - The n-th line of `Data.ref` must correspond to the n-th line of `Data.hyp` i.e., reference for n-th speech file, and associated speech recognition output
- Example
  - `Data.ref`  
eight four five  
five four seven  
zero seven six  
oh one three  
four five one  
...
  - `Data.hyp`  
eight eight four five  
five four seven  
zero seven six  
five oh one three  
four five one  
...



# Computation of word error rate (2)

Use the python module `wer.py`

```
wer -i Data.ref Data.hyp
```

Example of result

REF: \*\*\*\*\* eight four five

HYP: EIGHT eight four five

SENTENCE 1

Correct	= 100.0%	3	(	3)
---------	----------	---	---	----

Errors	= 33.3%	1	(	3)
--------	---------	---	---	----

REF: five four seven

HYP: five four seven

SENTENCE 2

Correct	= 100.0%	3	(	3)
---------	----------	---	---	----

Errors	= 0.0%	0	(	3)
--------	--------	---	---	----

REF: zero seven six

HYP: zero seven six

SENTENCE 3

Correct	= 100.0%	3	(	3)
---------	----------	---	---	----

Errors	= 0.0%	0	(	3)
--------	--------	---	---	----

REF: \*\*\*\*\* oh one three

HYP: FIVE oh one three

SENTENCE 4

Correct	= 100.0%	3	(	3)
---------	----------	---	---	----

Errors	= 33.3%	1	(	3)
--------	---------	---	---	----

# Estimated word error rate and confidence interval

The end of the file produced by the module `wer` looks like

Sentence count: 40

<b>WER:</b>	<b>19.167%</b> (	<b>23 /</b>	<b>120)</b>
WRR:	99.167% (	119 /	120)
SER:	100.000% (	40 /	40)

Word error rate

Number of errors

Number of word occurrences

**Confidence intervals** ⇔ uncertainty on the estimated word error rate

95% confidence interval :

$$1.96 \sqrt{\frac{P \cdot (1 - P)}{N}}$$

where  $P$  is the word error rate, and  $N$  the number of words in the test set

Here:

$$1.96 \sqrt{\frac{0.19 \cdot (1 - 0.19)}{120}} = 0.070 \text{ that is a word error rate of } 19.2\% \pm 7.0\%$$

# Preparing lexicons and language models

- Lexicon corresponding to English digits, including 'oh'
  - Lexicon: *zero, one, ..., nine, oh*
  - Extract words and associated pronunciations from the pocketsphinx lexicon (don't forget pronunciation variants)
- Digit loop jsgf grammar
  - Grammar allowing sequences of digits of any length
- Jsgf grammar for digit sequences of known length
  - 3 entry points corresponding to sequences of 1 digit, 3 digits and 5 digits
  - Can be specified in the same jsgf grammar file

# Adaptation of program examples for the experiments

1. **Adapt** the programs to handle the different jsfg format digit grammars
2. 3 programs **to write and test** on one or several speech files (digit sequences)
  - One using generic ngram language model
  - One using the digit loop grammar
  - One using a digit sequence of known length (of course, matching the length of the digit sequence to recognize)
3. **The make another modification to handle a set of speech files, and to write the speech recognition results in an output text file, for example**

```
temp/digits/raw/SNR05dB/boy/seq3digits_40_files/SNR05dB_boy_seq3digits_040.raw :: five
temp/digits/raw/SNR05dB/boy/seq3digits_40_files/SNR05dB_boy_seq3digits_022.raw :: oh five eight eight
temp/digits/raw/SNR05dB/boy/seq3digits_40_files/SNR05dB_boy_seq3digits_023.raw :: zero eight oh three oh
temp/digits/raw/SNR05dB/boy/seq3digits_40_files/SNR05dB_boy_seq3digits_024.raw :: one nine three five
temp/digits/raw/SNR05dB/boy/seq3digits_40_files/SNR05dB_boy_seq3digits_025.raw :: nine eight six
```

# Estimating word error rates

1. Make a script (program) that reads an output text file, and creates
  1. A reference file « `Data.ref` » by getting the reference transcript of each speech file
  2. An ASR hypothesis file « `Data.hyp` » that contain the speech recognition results
2. Apply the program computing the speech recognition errors

```
wer -i Data.ref Data.hyp > Data.results
```
3. Look at the content of the output file « `Data.results` »
4. Extract the word error rate on the data set,  
and **compute** the corresponding 95% confidence interval

# 1 – Impact of the language model

- Data
  - SNR35dB  $\Leftrightarrow$  very low noise level
  - man  $\Leftrightarrow$  man speakers
  - 1 digit, 3 digits & 5 digits  $\Leftrightarrow$  i.e., 400 files
- Language models
  - Generic ngram language model
  - Digit loop grammar in jsgf format
  - Digit sequence of known length (1 digit or 3 digits or 5 digits) in jsgf format
- Evaluation to do
  - **For each language model**, compute the word error rate and the associated 95% confidence interval on the 400 speech files corresponding to man speakers and very low noise level (35 dB SNR)

# 2 – Variability with respect to speaker groups

- Data
  - SNR35dB  $\Leftrightarrow$  very low noise level
  - **man, woman, boy, girl**  $\Leftrightarrow$  i.e., all the four speaker groups
  - 1 digit, 3 digits & 5 digits  $\Leftrightarrow$  i.e., 400 speech files per speaker group
- Language model
  - Jsgf grammar corresponding to sequences of digits of know length (1 digit or 3 digits or 5 digits)
- Evaluation to do
  - **For each speaker group** (*man, woman, boy, girl*), compute the word error rate and the associated 95% confidence interval on the 400 speech files corresponding to each speaker group, and very low noise level (35 dB SNR)

# 3 – Performance with respect to the length of the digit sequence

- Data
  - SNR35dB  $\Leftrightarrow$  very low noise level
  - man, woman  $\Leftrightarrow$  i.e., only adult speech data
  - 1 digit, 3 digits & 5 digits  $\Leftrightarrow$  i.e., 400 files per speaker group
- Language model
  - Jsgf grammar corresponding to sequences of digits of know length (1 digit or 3 digits or 5 digits)
- Evaluation to do
  - **For each length of digit sequences** (*1 digit, 3 digits, 5 digits*), compute the word error rate and the associated 95% confidence interval on the speech files corresponding to each category (400 files for 1 digit, 200 files for 3 & 5 digit sequences), and very low noise level (35 dB SNR)



# 4 – Impact of the noise level

- Data
  - SNR35dB, SNR25dB, SNR15dB & SNR05dB
  - man  $\Leftrightarrow$  i.e., only man speakers
  - 1 digit, 3 digits & 5 digits  $\Leftrightarrow$  i.e., 400 speech files
- Language model
  - Jsgf grammar corresponding to sequences of digits of known length (1 digit or 3 digits or 5 digits)
- Evaluation to do
  - **For each signal to noise ratio** (SNR35dB, SNR25dB, SNR15dB & SNR05dB), compute the word error rate and the associated 95% confidence interval on the 400 speech files corresponding to each SNR category

# Data and results to return

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(Preferably by uploading zip files on ARCHE (UL ENT web site), or else by mail, if any problem)

- Per group (if you work by groups of at most two people), or individually
  - The programs you have written and used, as well as grammar files and lexicons
  - The program output corresponding to the various speech recognition experiments, and to the computation of the word error rates (output of program `wer`)
  - A short document (2 to 3 pages) summarizing the experiments, and presenting and discussing the results (WERs of the four previous experiments)